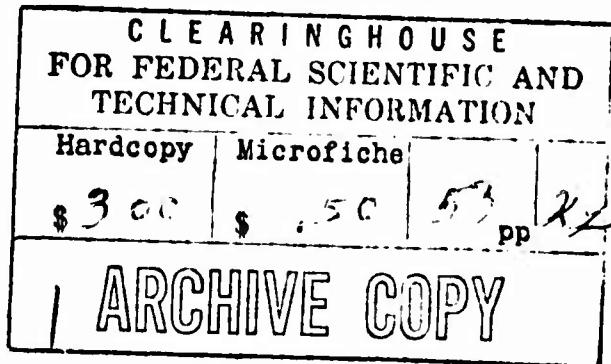


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COMPUTER ANALYSIS OF THREE-FACTOR
INTERACTIONS IN CONTINGENCY TABLES

Marvin A. Kastenbaum and Dennis Kuba

MRC Technical Summary Report #636
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ABSTRACT

Numerous methods have been proposed for testing the hypothesis of zero three-factor interaction in contingency tables. These methods are incorporated into a single computer program which is designed to calculate the test statistic by any or all of the proposed techniques, and which also provides estimators of the interactions and corresponding two-sided confidence intervals. The program is written in FORTRAN 63 for the CDC-3600 computer, and will analyze interactions in contingency tables of dimensions $2 \leq r \leq 5$, $2 \leq s \leq 5$, $2 \leq t \leq 16$, for $r \leq s \leq t$.

COMPUTER ANALYSIS OF THREE-FACTOR INTERACTIONS
IN CONTINGENCY TABLES

Marvin A. Kastenbaum and Dennis Kuba

1. Introduction

The analysis of three-factor interactions in contingency tables has been the subject of numerous recent papers culminating with the admirably lucid treatment and summary given by Goodman [15]. This series of papers begins with one by Kastenbaum and Lamphiear [8] in which the authors present an iterative technique for solving $(r-1)(s-1)(t-1)$ simultaneous fourth-degree equations. These equations result from an extension Bartlett's [2] test of zero interaction in a $2 \times 2 \times 2$ table to the general three-way ($r \times s \times t$) table as presented by Roy and Kastenbaum [6]. Subsequent authors [10, 11, 12, 13, 14, 15] have proposed alternative techniques of analysis, each progressively simpler than the preceding one, but all testing the same hypothesis. In every case; the test statistic is distributed asymptotically as chi-square with $(r-1)(s-1)(t-1)$ degrees of freedom.

It is the purpose of this paper to describe a computer program which has been designed to calculate the test statistic by any or all of the proposed techniques. This program also provides the estimates of interaction and corresponding simultaneous confidence intervals given by Goodman [15]. The program was written in FORTRAN 63 for the CDC-3600 computer, and will do analyses on contingency tables of dimensions $2 \leq r \leq 5$, $2 \leq s \leq 5$, $2 \leq t \leq 16$, for $r \leq s \leq t$. [Appendix].

2. Tests of Hypotheses

Let n_{ijk} be the number of observations in the i th row, j th column, and

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kth layer of a three-way contingency such that $\sum_{i=1}^r \sum_{j=1}^s \sum_{k=1}^t n_{ijk} = n$, the total sample size. Denote by $0 < \Pi_{ijk} < 1$ the corresponding probability where

$\sum_{i=1}^r \sum_{j=1}^s \sum_{k=1}^t \Pi_{ijk} = 1$. Also define the following relationships:

$$\sum_{i=1}^r n_{ijk} = n_{\cdot jk}, \sum_{j=1}^s n_{ijk} = n_{i \cdot k}, \sum_{k=1}^t n_{ijk} = n_{ij \cdot},$$

$$\sum_{i=1}^r \sum_{j=1}^s n_{ijk} = n_{\cdot \cdot k}, \sum_{i=1}^r \sum_{k=1}^t n_{ijk} = n_{i \cdot \cdot}, \sum_{j=1}^s \sum_{k=1}^t n_{ijk} = n_{\cdot \cdot \cdot},$$

with similar relationships for the Π_{ijk} 's. The hypothesis of zero three-factor interaction in an $r \times s \times t$ contingency table is given by:

$$H_0: \frac{\Pi_{rsk} \Pi_{ijk}}{\Pi_{isk} \Pi_{rjk}} = \frac{\Pi_{rst} \Pi_{ijt}}{\Pi_{ist} \Pi_{rjt}} \quad (2.1)$$

for all integers i, j, k such that $1 \leq i < r, 1 \leq j < s, 1 \leq k < t$.

2.1. Kastenbaum and Lamphiear [8]

For the $r \times s \times t$ table, Kastenbaum and Lamphiear generalized an iterative technique proposed by Norton [4] for handling the simultaneous fourth-degree equations which arise in the estimation process. Under the null hypothesis (2.1), estimates of the parameters may be achieved by first solving for all x_{ijk} in the following systems of equations:

$$\frac{(n_{rsk} - \sum_{i=1}^{r-1} \sum_{j=1}^{s-1} x_{ijk})(n_{ijk} - x_{ijk})}{(n_{isk} + \sum_{j=1}^{s-1} x_{ijk})(n_{rjk} + \sum_{i=1}^{r-1} x_{ijk})} = \frac{(n_{rst} - \sum_{i=1}^{r-1} \sum_{j=1}^{s-1} x_{ijt})(n_{ijt} - x_{ijt})}{(n_{ist} + \sum_{j=1}^{s-1} x_{ijt})(n_{rjt} + \sum_{i=1}^{r-1} x_{ijt})}, \quad (2.1.1)$$

for all positive integers $i < r, j < s$, and $k < t$, and subject to the constraints

$\sum_{k=1}^{t-1} x_{ijk} = -x_{ijt}$ for all positive integers $i \leq r$ and $j \leq s$. Let v_{ijk} be the values of x_{ijk} which satisfy (2.1.1) for all integers $1 \leq i \leq r$, $i \leq j \leq s$, and $1 \leq k \leq t$. Then

$$X^2 = \sum_{i=1}^r \sum_{j=1}^s \sum_{k=1}^t v_{ijk}^2 / (n_{ijk} - v_{ijk}) \quad (2.1.2)$$

is distributed asymptotically as chi-square with $(r-1)(s-1)(t-1)$ degrees of freedom.

The program solves equations (2.1.1) by applying Newton's method of functional iteration to a technique proposed by Norton [4], and computes the value of the test statistic (2.1.2). It then prints out the number of iterations, the number of degrees of freedom, the value of the test statistic, and the observed and expected cell frequencies, with identification for all the cells in the contingency table. A sample print-out is given in Table 1.

2.2. Darroch [10]

An alternative solution to equations (2.1.1) is given by Darroch. This method involves the iterative solution for δ_{jk} , φ_{ki} , and ψ_{ij} in the following $(rs + rt + st)$ simultaneous non-linear equations:

$$\frac{n_{\cdot jk}}{n} = \delta_{jk} \sum_{i=1}^r \varphi_{ki} \psi_{ij} \quad (2.2.1)$$

$$\frac{n_{i \cdot k}}{n} = \varphi_{ki} \sum_{j=1}^s \psi_{ij} \delta_{jk} \quad (2.2.2)$$

$$\frac{n_{ij\cdot}}{n} = \psi_{ij} \sum_{k=1}^t \delta_{jk} \varphi_{ki} \quad (2.2.3)$$

The iteration begins by setting $\varphi_{ki} = \varphi_{ki}^{(1)} = n_{i \cdot k} / n_{\cdot \cdot k}$ and $\psi_{ij} = \psi_{ij}^{(1)} = n_{ij\cdot} / n_{i \cdot \cdot}$ in equation (2.2.1) and solving for $\delta_{jk} = \delta_{jk}^{(2)}$, then in equation (2.2.2) let

$\psi_{ij} = \psi_{ij}^{(1)}$ and $\delta_{jk} = \delta_{jk}^{(2)}$ and solve for $\phi_{ki} = \phi_{ki}^{(2)}$. In equation (2.2.3), let $\delta_{jk} = \delta_{jk}^{(2)}$ and $\phi_{ki} = \phi_{ki}^{(2)}$ and solve for $\psi_{ij} = \psi_{ij}^{(2)}$. Then return to equation (2.2.1), letting $\phi_{ki} = \phi_{ki}^{(2)}$ and $\psi_{ij} = \psi_{ij}^{(2)}$, and solve for $\delta_{jk} = \delta_{jk}^{(3)}$. Continue the iteration in this way until $\psi_{ij}^{(m)} = \psi_{ij}^{(m-1)}$, $\delta_{jk}^{(m)} = \delta_{jk}^{(m-1)}$, and $\phi_{ki}^{(m)} = \phi_{ki}^{(m-1)}$ to five decimal places for all positive integers $i \leq r$, $j \leq s$, $k \leq t$.

At this point calculate

$$x^2 = \sum_{i=1}^r \sum_{j=1}^s \sum_{k=1}^t [n_{ijk} - n \delta_{jk}^{(m)} \phi_{ki}^{(m)} \psi_{ij}^{(m)}]^2 / n \delta_{jk}^{(m)} \phi_{ki}^{(m)} \psi_{ij}^{(m)}. \quad (2.2.4)$$

Darroch has shown that this test statistic is distributed asymptotically as chi-square with $(r-1)(s-1)(t-1)$ degrees of freedom. Moreover (2.2.4) and (2.1.2) are algebraically identical.

The program solves equations (2.2.1), (2.2.2), and (2.2.3), and computes the value of the test statistic (2.2.4). The print-out is identical to that of Table 1, except that the number of iterations is generally smaller. In this sense, the Darroch procedure is superior to the Kastenbaum-Lamphiear procedure.

2.3. Plackett [11]

Let $\theta_{ijk} = \Pi_{ijk} / \Pi_{...k}$ be the conditional probability that an observation will fall in the i th row and j th column, given that it is in the k th layer. It follows that the null hypothesis, (2.1), may be rewritten as

$$H_0: \Delta_{ijk} = \Delta_{ijt} \quad (2.3.1)$$

for all integers $1 \leq i < r$, $1 \leq k < t$, where $\Delta_{ijk} = \theta_{rsk} \theta_{ijk} / \theta_{isk} \theta_{rjk}$. Using a criterion suggested by Woolf [5], Plackett presents a procedure for testing a null hypothesis equivalent to (2.3.1), namely

$$H_0: \Gamma_{ijk} = \Gamma_{ijt} \quad (2.3.2)$$

for all integers $1 \leq i < r$, $1 \leq j < s$, $1 \leq k < t$, where $\Gamma_{ijk} = \log \Delta_{ijk}$. This procedure is based on the fact that the maximum likelihood estimator of $\log \Delta_{ijk}$ is $\log d_{ijk}$, where $d_{ijk} = n_{ijk} n_{rsk} / n_{isk} n_{rjk}$ for all integers $1 \leq i < r$, $1 \leq j < s$, $1 \leq k < t$. Moreover the variance of $\log d_{ijk}$ may be estimated consistently by

$$u_{ijk} = \frac{1}{n_{ijk}} + \frac{1}{n_{rsk}} + \frac{1}{n_{isk}} + \frac{1}{n_{rjk}} .$$

If R and S are two matrices of order $(r-1) \times r$ and $(s-1) \times s$ respectively with rows orthogonal to each other and to the unit vector, then the direct product, $[R * S]$, is a matrix with $(r-1)(s-1)$ rows and rs columns. The elements of each row of this matrix provide the coefficients of a linear combination of the logarithm of the frequencies in the kth layer of the contingency table. More specifically, the matrices R and S are formed as follows:

$$R = \{\rho_{\alpha i}\} = \begin{cases} 1 & \text{for } 1 \leq i \leq \alpha, \\ -\alpha & \text{for } i = \alpha + 1, \\ 0 & \text{for } i > \alpha + 1, \text{ where } 1 \leq i \leq r, 1 \leq \alpha < r, \text{ and} \end{cases}$$

$$S = \{\sigma_{\beta j}\} = \begin{cases} 1 & \text{for } 1 \leq j \leq \beta, \\ -\beta & \text{for } j = \beta + 1, \\ 0 & \text{for } j > \beta + 1, \text{ where } 1 \leq j \leq s, 1 \leq \beta < s. \end{cases}$$

Then for each positive integer $k \leq t$, a column vector z_k , is generated from the product of the $(r-1)(s-1) \times rs$ matrix $[R * S]$ with the $rs \times 1$ vector $\{\log n_{ijk}\}$.

The elements of z_k are

$$z_{k\alpha\beta} = \sum_{i=1}^r \sum_{j=1}^s \rho_{\alpha i} \sigma_{\beta j} \log n_{ijk} . \quad (2.3.3.)$$

The asymptotic distribution of $z_k = \{z_{k\alpha\beta}\}$ is multivariate normal, with dispersion matrix

$$V_k = [R * S] D_{ijk}^{-1} [R * S]^T ,$$

where $D_{n_{ijk}}^{-1}$ is a square matrix of order rs with elements $\{\frac{1}{n_{ijk}}\}$ on the diagonal and zeros elsewhere. The elements of V_k are

$$\text{Cov}[z_{k\alpha\beta}, z_{k\alpha'\beta'}] = \sum_{i=1}^r \sum_{j=1}^s \rho_{\alpha i} \rho_{\alpha' i} \sigma_{\beta j} \sigma_{\beta' j} / n_{ijk}, \quad (2.3.4)$$

for all positive integers $\alpha, \alpha' < r$ and $\beta, \beta' < s$. If z'_k denotes the transpose of z_k , and V_k^{-1} the inverse of V_k , then on the hypothesis of zero three-factor interaction, (2.3.2), the statistic

$$Y^2 = \sum_{k=1}^t z'_k V_k^{-1} z_k - [\sum_{k=1}^t z'_k V_k^{-1}] [\sum_{k=1}^t V_k^{-1}]^{-1} [\sum_{k=1}^t V_k^{-1} z_k] \quad (2.3.5)$$

is distributed asymptotically as chi-square with $(r-1)(s-1)(t-1)$ degrees of freedom.

This portion of the program computes the elements of all the vectors z_k and of their associated dispersion matrices V_k from formulas (2.3.3) and (2.3.4) for all integers $1 \leq k \leq t$. To evaluate the test statistic, (2.3.5), $(t+1)$ square matrices of order $(r-1)(s-1)$ are inverted. The print-out, Table 2, displays all the vectors z_k and their associated dispersion matrices V_k for all integers $1 \leq k \leq t$; and $\sum_{k=1}^t L_k^2 = \sum_{k=1}^t z'_k V_k^{-1} z_k$, $P = [\sum_{k=1}^t V_k^{-1}]^{-1}$, $\sum_{k=1}^t z'_k V_k^{-1}$, as well as the value of the test statistic, (2.3.5), and the number of degrees of freedom.

2.4. Goodman [14, 15]

The approach suggested by Goodman modifies Plackett's method for testing the null hypothesis (2.3.2) by redefining the two matrices R and S as follows:

$$R = \{\rho_{\alpha i}\} = \begin{cases} 1 & \text{for } i = \alpha, \\ -1 & \text{for } i = r, \\ 0 & \text{elsewhere, where } 1 \leq i \leq r, 1 \leq \alpha < r, \text{ and} \end{cases}$$

$$S = \{\sigma_{\beta j}\} = \begin{cases} 1 & \text{for } j = \beta \\ -1 & \text{for } j = s \\ 0 & \text{elsewhere, where } 1 \leq j \leq s, 1 \leq \beta < s \end{cases} .$$

As in Plackett's method, column vectors $g_k = \{g_{k\alpha\beta}\}$ are generated, where

$$g_{k\alpha\beta} = \sum_{i=1}^r \sum_{j=1}^s \rho_{\alpha i} \sigma_{\beta j} \log n_{ijk} . \quad (2.4.1)$$

The asymptotic distribution of each g_k is multivariate normal with dispersion matrix U_k , whose elements are

$$\text{Cov}[g_{k\alpha\beta}, g_{k\alpha'\beta'}] = \sum_{i=1}^r \sum_{j=1}^s \rho_{\alpha i} \rho_{\alpha' i} \sigma_{\beta j} \sigma_{\beta' j} / n_{ijk} \quad (2.4.2)$$

for all positive integers $\alpha, \alpha' < s$. If $U_k^{-1} = M_k$, $Q = [\sum_{k=1}^t M_k]^{-1}$, and $\tilde{g} = \sum_{k=1}^t M_k g_k$, then the statistic

$$Y^2 = \sum_{k=1}^t g_k' M_k g_k - \tilde{g}' Q \tilde{g} \quad (2.4.3)$$

is distributed asymptotically as chi-square with $(r-1)(s-1)(t-1)$ degrees of freedom. Moreover, expressions (2.3.5) and (2.4.3) are identical.

Except for the new definitions of the matrices R and S , the computational procedure using either Plackett's or Goodman's methods appear to be identical. In both methods it is necessary to invert $(t+1)$ matrices of order $(r-1)(s-1)$. However, Goodman makes a significant contribution, at this point, by demonstrating that only one $(r-1)(s-1)$ matrix and t matrices of order $(r-1)$, for $r \leq s$, need to be inverted in calculating the test statistic (2.4.3). The reduction in computing time which results from this modified procedure may be appreciable. The calculations are carried out as follows:

(i) For every integer k , $1 \leq k \leq t$, define the $(r-1) \times (r-1)$ matrices

$$B^{(k)} = D_{n_{isk}} - \frac{1}{n_{.sk}} \{n_{isk}\} \{n_{isk}\}' ,$$

$$B_j^{(k)} = D_{n_{ijk}} - \frac{1}{n_{.jk}} \{n_{ijk}\} \{n_{ijk}\}' ,$$

where $D_{n_{isk}}$ and $D_{n_{ijk}}$ are $(r-1) \times (r-1)$ diagonal matrices with elements n_{isk} and n_{ijk} respectively, and where $\{n_{isk}\}'$ and $\{n_{ijk}\}'$ are the respective transposes of the $(r-1) \times 1$ column vectors $\{n_{isk}\}$ and $\{n_{ijk}\}$, for integers $1 \leq i < r$, and $1 \leq j < s$.

(ii) Evaluate $C^{(k)} = B^{(k)} + \sum_{j=1}^{s-1} B_j^{(k)}$, an $(r-1) \times (r-1)$ matrix, and its inverse $G^{(k)}$, $1 \leq k \leq t$.

(iii) Construct the submatrices

$$M_{jj'}^{(k)} = \begin{cases} B_j^{(k)} - B_j^{(k)} G^{(k)} B_j^{(k)} & \text{for } j = j' = 1, 2, \dots, s-1 , \\ - B_j^{(k)} G^{(k)} B_{j'}^{(k)} & \text{for } j \neq j' . \end{cases}$$

(iv) For every k , $1 \leq k \leq t$, form the square matrix M_k of order $(r-1)(s-1)$ in equation (2.4.3) from the $(s-1)^2$ submatrices $M_{jj'}^{(k)}$ each of order $(r-1) \times (r-1)$.

(v) Evaluate the test statistic using equation (2.4.3).

The print-out (Table 3), for this section of the program displays all the vectors g_k , their associated dispersion matrices U_k , and $\sum_{k=1}^t H_k^2 = \sum_{k=1}^t g_k' M_k g_k$, Q , g' , as well as the value of the test statistic, (2.4.3), and the number of degrees of freedom.

2.5. Goodman [15]: The $2 \times 2 \times t$ Contingency Table

For the three-way contingency table with $r = 2$ rows, $s = 2$ columns, and $t \geq 2$ layers, Goodman proposes three alternative test statistics. Two of these are based on an analysis of the cell frequencies, and the third, which is a special case of the procedure discussed in section 2.4, is based on an analysis of the log-frequencies.

$$(i) \text{ Define } d_k = \frac{n_{11k} n_{22k}}{n_{12k} n_{21k}}, \quad u_k = \frac{1}{n_{11k}} + \frac{1}{n_{22k}} + \frac{1}{n_{12k}} + \frac{1}{n_{21k}}, \quad v_k = d_k^2 u_k,$$

and $w_k = 1/v_k$, $1 \leq k \leq t$. Then

$$x^2 = \sum_{k=1}^t d_k^2 w_k - \left[\sum_{k=1}^t d_k w_k \right]^2 / \sum_{k=1}^t w_k \quad (2.5.1)$$

is distributed asymptotically as chi-square with $(t-1)$ degrees of freedom.

Equation (2.5.1) may be used to test the hypothesis of zero three-factor interaction which, in the $2 \times 2 \times t$ table may be specified as

$$H_0: \frac{\pi_{111} \pi_{221}}{\pi_{121} \pi_{211}} = \frac{\pi_{11k} \pi_{22k}}{\pi_{12k} \pi_{21k}} \quad 2 \leq k \leq t. \quad (2.5.2)$$

Equation (2.5.2) may be rewritten as

$$H_0: \Delta_1 = \Delta_k, \quad 2 \leq k \leq t, \quad (2.5.3)$$

where $\Delta_k = \frac{\pi_{11k} \pi_{22k}}{\pi_{12k} \pi_{21k}} = \frac{\theta_{11k} \theta_{22k}}{\theta_{12k} \theta_{21k}}$

is a measure of the two factor interaction in the k th layer of the table. The maximum likelihood estimator of Δ_k is d_k , and its variance can be estimated consistently by v_k .

Moreover, Goodman points out that this null hypothesis may be partitioned into the following $(t-1)$ sub-hypotheses:

$$H_1: \Delta_1 = \Delta_2$$

$$H_2: \Delta_1 = \Delta_3 \text{ (given that } \Delta_1 = \Delta_2)$$

$$H_3: \Delta_1 = \Delta_4 \text{ (given that } \Delta_1 = \Delta_2 = \Delta_3) \quad (2.5.4)$$

⋮

$$H_{t-1}: \Delta_1 = \Delta_t \text{ (given that } \Delta_1 = \Delta_2 = \dots = \Delta_{t-1}) .$$

Each of these sub-hypotheses H_k ($1 \leq k < t$) can be tested using a single degree of freedom as follows:

$$\text{Define } d_\gamma = \frac{n_{11\gamma} n_{22\gamma}}{n_{12\gamma} n_{21\gamma}} \quad 1 \leq \gamma < k ,$$

and corresponding quantities u_γ , v_γ , and w_γ . Let $d_k^* = \sum_{\gamma=1}^k d_\gamma w_\gamma / \sum_{\gamma=1}^k w_\gamma$ and $v_k^* = 1 / \sum_{\gamma=1}^k w_\gamma$. Then to test H_k , $1 \leq k < t$, in equation (2.5.4) compute

$$x_k^2 = [d_{k+1} - d_k^*]^2 / [v_{k+1} + v_k^*] . \quad (2.5.5)$$

The sum, $\sum_{k=1}^{t-1} x_k^2$ is equal to x^2 given by (2.5.1).

When H_k is true, (2.5.5) is distributed asymptotically as chi-square with one degree of freedom. Also when H_0 is true, each of the $(k-1)$ statistics x_k^2 will have asymptotically independent chi-square distributions each with one degree of freedom.

$$(ii) \text{ Let } b_k = \frac{1}{d_k}, \quad a_k = b_k^2 u_k, \quad h_k = \frac{1}{a_k}, \quad 1 \leq k \leq t ,$$

$$b_\gamma = \frac{1}{d_\gamma}, \quad a_\gamma = b_\gamma^2 u_\gamma, \quad h_\gamma = \frac{1}{a_\gamma}, \quad 1 \leq \gamma \leq k ,$$

$$b_k^* = \sum_{\gamma=1}^k b_\gamma h_\gamma / \sum_{\gamma=1}^k h_\gamma, \quad \text{and} \quad a_k^* = 1 / \sum_{\gamma=1}^k h_\gamma .$$

Here b_k is the maximum likelihood estimate of $1/\Delta_k$, and its variance can be estimated consistently by a_k . When the null hypothesis (2.5.3) is true

$$Z^2 = \sum_{k=1}^t b_k^2 h_k - [\sum_{k=1}^t b_k h_k]^2 / \sum_{k=1}^t h_k \quad (2.5.6)$$

is distributed asymptotically as chi-square with $(t-1)$ degrees of freedom. Tests of H_k ($1 \leq k < t$) in equation (2.5.4) are given by

$$z_k^2 = [b_{k+1} - b_k^*]^2 / [a_{k+1} + a_k^*] \quad (2.5.7)$$

which, when H_k is true, is distributed asymptotically as chi-square with one degree of freedom. Also $\sum_{k=1}^{t-1} z_k^2 = Z^2$.

(iii) Let $g_k = \log d_k$, $m_k = 1/u_k$, $1 \leq k \leq t$,

$$g_\gamma = \log d_\gamma, \quad m_\gamma = 1/u_\gamma, \quad 1 \leq \gamma \leq k,$$

$$g_k^* = \sum_{\gamma=1}^k g_\gamma m_\gamma / \sum_{\gamma=1}^k m_\gamma, \quad \text{and} \quad u_k^* = 1 / \sum_{\gamma=1}^k m_\gamma.$$

Here g_k is the maximum likelihood estimate of $\Gamma_k = \log \Delta_k$, and its variance can be estimated consistently by u_k . The null hypothesis, (2.5.3), becomes

$$H_0: \Gamma_1 = \Gamma_k \quad \text{for} \quad 2 \leq k \leq t. \quad (2.5.8)$$

When (2.5.8) is true

$$Y^2 = \sum_{k=1}^t g_k^2 m_k - [\sum_{k=1}^t g_k m_k]^2 / \sum_{k=1}^t m_k \quad (2.5.9)$$

is distributed asymptotically as chi-square with $(t-1)$ degrees of freedom. This statistic, (2.5.9), is the special form which equation (2.4.3) takes when $r = 2$ and $s = 2$. Tests of hypotheses analogous to (2.5.4) involving the log-frequencies are given by

$$Y_k^2 = [g_{k+1} - g_k^*]^2 / [u_{k+1} + u_k^*] , \quad (2.5.10)$$

for every integer k , $1 \leq k < t$, which, when H_k is true, is distributed asymptotically as chi-square with one degree of freedom. Also, $\sum_{k=1}^t Y_k^2 = Y^2$. Moreover, when H_k is true, Y_k^2 and Z_k^2 are asymptotically equivalent to X_k^2 .

This portion of the program computes all the test statistics given by equations (2.5.1), (2.5.5), (2.5.6), (2.5.7), (2.5.9) and (2.5.10). The printout, displayed in Table 4, gives values of d_k , v_k , b_k , a_k , g_k , and u_k for $1 \leq k \leq t$, and X_k^2 , Z_k^2 , and Y_k^2 for $1 \leq k < t$. In addition, the program computes and displays the values X^2 , Z^2 , Y^2 , $\sum_{k=1}^{t-1} X_k^2$, $\sum_{k=1}^{t-1} Z_k^2$, $\sum_{k=1}^{t-1} Y_k^2$, and the number of degrees of freedom.

2.6. Kullback, Kupperman, and Ku [12,13]

The procedures outlined by Plackett [11] and Goodman [14,15] for testing the hypothesis of zero three-factor interaction, (2.1), are primarily generalizations of a method proposed by Woolf [5] in which the test criterion is based on a logit transformation of the data. This test criterion is one alternative to the Pearson [1] chi-square test of goodness of fit. Another alternative is the likelihood-ratio test criterion proposed by Wilks [3], and investigated by Woolf [7] and Kullback, Kupperman, and Ku [12,13].

Kullback, Kupperman, and Ku resort to an information theory approach and define a minimum discrimination information statistic [M.D.I.S.]. This statistic is distributed asymptotically as chi-square under the null hypothesis, and as noncentral chi-square under the alternative hypothesis, with appropriate degrees

of freedom and noncentrality parameter. It has additive properties in the sense that it can be analyzed into several additive components for a hypothesis which is equivalent to the combination of several hypotheses of interest. Each such component of the M. D. I. S. is itself an M. D. I. S., and is distributed asymptotically as chi-square with appropriate degrees of freedom. Moreover, the M. D. I. S. has the convexity property which is useful in finding other M. D. I. S. under certain restrictions and groupings. In particular, for nonnegative real numbers a_i and b_i , this property yields

$$\sum_{i=1}^n a_i \ln(a_i/b_i) \geq \left(\sum_{i=1}^n a_i \right) \ln\left(\sum_{i=1}^n a_i / \sum_{i=1}^n b_i\right) . \quad (2.6.1)$$

Equality holds if and only if $\frac{a_1}{b_1} = \frac{a_2}{b_2} = \dots = \frac{a_n}{b_n}$.

In the $r \times s \times t$ table, the criterion for testing H_0 , (2.1), is given by

$$2f = 2 \sum_{i=1}^r \sum_{j=1}^s \sum_{k=1}^t n_{ijk} \log \left[\frac{n_{ijk} n_{i..} n_{.j.} n_{...k}}{n_{ij.} n_{i..} n_{.jk} n_{...}} \right] . \quad (2.6.2)$$

This statistic is distributed asymptotically as chi-square with $(r-1)(s-1)(t-1)$ degrees of freedom. In any specific example the convexity property may not hold. This may result in a negative value for (2.6.2), in which case Kullback, Kupperman and Ku [12, pp. 225-226] recommend one of the following alternative analyses of the data:

row-layer interaction with column

$$2\hat{I} = 2 \sum_{i=1}^r \sum_{j=1}^s \sum_{k=1}^t n_{ijk} \log \left[\frac{n_{ijk} n_{..j.} y_{i..k}}{n_{ij.} n_{..jk} n_{i..k}} \right], \quad (2.6.3)$$

where $y_{i..k} = \sum_{j=1}^s n_{ij.} n_{..jk} / n_{..j.}$;

column-layer interaction with row

$$2\hat{I} = 2 \sum_{i=1}^r \sum_{j=1}^s \sum_{k=1}^t n_{ijk} \log \left[\frac{n_{ijk} n_{i..} y_{..jk}}{n_{ij.} n_{..jk} n_{i..k}} \right] \quad (2.6.4)$$

where $y_{..jk} = \sum_{i=1}^r n_{ij.} n_{i..k} / n_{i..}$;

row-column interaction with layer

$$2\hat{I} = 2 \sum_{i=1}^r \sum_{j=1}^s \sum_{k=1}^t n_{ijk} \log \left[\frac{n_{ijk} n_{..k} y_{ij.}}{n_{ij.} n_{..jk} n_{i..k}} \right] \quad (2.6.5)$$

where $y_{ij.} = \sum_{k=1}^t n_{i..k} n_{..jk} / n_{..k}$.

The program computes the M.D.I.S. test statistics using equations (2.6.2), (2.6.3), (2.6.4), and (2.6.5), and prints out the corresponding values of $2\hat{I}$ as well as the number of degrees of freedom.

3. Confidence Intervals

Of all the papers on the subject of three-factor interaction in contingency tables, only Goodman's [15] deals with the problem of estimation. Indeed, the final section of this paper is devoted exclusively to a discussion of methods for estimating the magnitude of the three-factor interaction, and for obtaining confidence limits for it.

Let

$$A_{ibjckd} = \log \left\{ \frac{\frac{\Pi_{ijk} \Pi_{bck}}{\Pi_{bjk} \Pi_{ick}}}{\frac{\Pi_{ijd} \Pi_{bcd}}{\Pi_{bjd} \Pi_{icd}}} \right\} \quad (3.1)$$

be the measure of a particular three-factor interaction for all integers $1 \leq i < b \leq r$, $1 \leq j < c \leq s$, and $1 \leq k < d \leq t$. Depending on the values of b , c , and d , there can be as many as $\binom{r}{2} \binom{s}{2} \binom{t}{2}$ such three-factor interactions in an $r \times s \times t$ contingency table. In particular, for $b = r$, $c = s$, $d = t$, there are $(r-1)(s-1)(t-1)$ interactions of the form

$$\begin{aligned} A_{irjskt} &= \Gamma_{ijk} - \Gamma_{ijt} = \log(\Delta_{ijk}/\Delta_{ijt}) \\ &= \log \left\{ \frac{\frac{\Pi_{ijk} \Pi_{rsk}}{\Pi_{rjk} \Pi_{isk}}}{\frac{\Pi_{ijt} \Pi_{rst}}{\Pi_{rjt} \Pi_{ist}}} \right\}. \end{aligned} \quad (3.2)$$

The maximum likelihood estimator of A_{ibjckd} is

$$a_{ibjckd} = \log \left\{ \frac{\frac{n_{ijk} n_{bck}}{n_{bjk} n_{ick}}}{\frac{n_{ijd} n_{bcd}}{n_{bjd} n_{icd}}} \right\}, \quad (3.3)$$

and its variance is estimated consistently by

$$S_{ibjckd}^2 = u_{ibjck} + u_{ibjcd}, \quad (3.4)$$

$$\text{where } u_{ibjck} = \frac{1}{n_{ijk}} + \frac{1}{n_{bck}} + \frac{1}{n_{ick}} + \frac{1}{n_{bjk}}. \quad (3.5)$$

Using these definitions, Goodman proposes three alternative sets of approximate two-sided confidence intervals for A_{ibjckd} :

$$a_{ibjckd} \pm \chi_T(P) S_{ibjckd}, \quad (3.6)$$

where $\chi_T^2(P)$ is the $[P \times 100]$ th percentile of the chi-square distribution with $T = (r-1)(s-1)(t-1)$ degrees of freedom;

$$a_{ibjckd} \pm \chi_1(P) S_{ibjckd} , \quad (3.7)$$

where $\chi_1(P)$ is the $[\frac{1+P}{2} \times 100]$ th percentile of the standardized normal distribution; and for a specified subset of interactions, $A_{(w)}$,

$$a_{(w)} \pm \Phi_W(P) S_{(w)} , \quad (3.8)$$

where, for any (b, c, d) , $w = (b-1)(c-1)(d-1)$ is the number of elements in the subset $(w) = (ibjckd)$; and

$$W = W(B, C, D) = \sum_{b \in B} \sum_{c \in C} \sum_{d \in D} (b-1)(c-1)(d-1) ,$$

where B, C , and D are subsets of the sets of integers I_r, I_s, I_t such that

$$B \subseteq I_r = \{2, 3, \dots, r\} ,$$

$$C \subseteq I_s = \{2, 3, \dots, s\} ,$$

$$D \subseteq I_t = \{2, 3, \dots, t\} ;$$

and $\Phi_W(P)$ is the $(\frac{2W-1+P}{2W} \times 100)$ percentile of the standardized normal distribution. It follows that $1 \leq w \leq W \leq (\frac{r}{2})(\frac{s}{2})(\frac{t}{2})$. In particular, for the subset $(w) = (irjskt)$, $w = W = T = (r-1)(s-1)(t-1)$, and the intervals (3.8) become

$$a_{irjskt} \pm \Phi_T(P) S_{irjskt} , \quad (3.9)$$

where $\Phi_T(P)$ is the $(\frac{2T-1+P}{2T} \times 100)$ th percentile of the standardized normal distribution.

In the final paragraph of his paper, Goodman [15] discusses the relative merits of the three alternative sets of confidence intervals (3.6), (3.7), and (3.8). His remarks may be summarized as follows:

(i) If any of the intervals (3.6) do not include zero, the null hypothesis, (2.3.2), will be rejected at a significance level $(1-P)$, when tested using (2.4.3).

(ii) When a specific set of W three-factor interactions is of interest, $W \leq \binom{r}{2} \binom{s}{2} \binom{t}{2}$, $T \geq 1$, and when the usual values of P , [$P = .95$ or $.99$], are used, (3.8) will yield smaller confidence intervals than (3.6).

(iii) The probability is (approximately) at least P that all the $\binom{r}{2} \binom{s}{2} \binom{t}{2}$ intervals (3.6) include the corresponding true values. That is to say, the probability is (approximately) at least P that all the $\binom{r}{2} \binom{s}{2} \binom{t}{2}$ confidence statements associated with (3.6) are correct.

(iv) The length of the confidence interval will be reduced if (3.7) is used in place of (3.6). However, the probability that all the confidence statements are correct will also be reduced. The expected proportion of correct confidence statements of type (3.7) will be approximately P .

(v) To insure that all confidence statements are correct, with probability (approximately) at least P , use form (3.6). If it suffices to insure that the expected proportion of correct confidence statements is P , then (3.7) should replace (3.6).

(vi) The same consequences will result when intervals (3.7) are used in place of intervals (3.8) .

For a prescribed set of integers (b, c, d) , and for all (i, j, k) in the range $1 \leq i < b \leq r$, $1 \leq j < c \leq s$, $1 \leq k < d \leq t$, this portion of the program calculates the estimates of interaction and their estimated variances using equations (3.3) and (3.4). Moreover, it evaluates the three alternative sets of approximate two-sided confidence intervals given by equations (3.6), (3.7), and (3.8), for values of $P = 0.95$ and 0.99 . Values of $\chi^2_T(P)$ in (3.6) are calculated using the Fisher-Cornish [9] approximation.

Finally, the program is designed to transform the estimates of interaction and the corresponding confidence intervals to their original scale by taking the anti-logarithm of each of the computed values. The results of these calculations are displayed in logarithmic units in Table 5.

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Appendix

Program HYPOINT

```
PROGRAM HYPOINT
COMMON/ALL/UBS(400),IDENT(10),IR,JS,KT,NDEGREES,INPUT,OUTPUT
COMMON/FLAGS/OPTION(10),LOGFLAG
TYPE INTEGER OPTION
TYPE INTEGER OUTPUT
INPUT=60
OUTPUT=61
10 CALL READIN
   IF(OPTION(1).EQ.0) GO TO 20
   CALL KASTENBM
20 IF(OPTION(2)+OPTION(3).EQ.0) GO TO 30
   CALL DARROCH
30 IF(OPTION(4).EQ.0) GO TO 40
   CALL PLACKETT
40 IF(OPTION(5).EQ.0) GO TO 50
   CALL GOODMAN1
50 IF(OPTION(6).EQ.0) GO TO 60
   CALL GOODMAN2
60 IF(OPTION(7).EQ.0) GO TO 70
   CALL TWOBYTWO
70 IF(OPTION(8).EQ.0) GO TO 80
   CALL KKK
80 IF(OPTION(9).EQ.0) GO TO 90
   LOGFLAG=1
   CALL INTRVALS
90 IF(OPTION(10).EQ.0) GO TO 100
   LOGFLAG=0
   CALL INTRVALS
100 GO TO 10
END
```

```

SUBROUTINE KASTENBM
COMMON/ALL/OBS(400),IDENT(10),IR,JS,KT,NDEGREES,INPUT,OUTPUT
DIMENSION EXP(400),BIJ(16),CIJ(16)
TYPE INTEGER OUTPUT
2000 FORMAT(1H1,10A8)
2001 FORMAT(32HOKASTENBAUM-LAMPHIEAK PROCEDURE )
2002 FORMAT(////,12H ITERATION ,8X,18HDEGREES OF FREEDOM,8X,
*      14H     CHI-SQUARE,15X,2H R,8X,2H S,8X,2H T)
2003 FORMAT(////////,8X,15H     CELL   ,8X,18HOBERVED FREQUENCY,8X,
*      18HEXPECTED FREQUENCY /,1H )
2004 FORMAT(/3X,I4,18X,I5,15X,F15.6,7X,3(8X,I2))
2005 FORMAT(6X,3(2X,I3),8X,F15.6,11X,F15.6)
2006 FORMAT(25H0ITERATION COUNT EXCEEDS I3)
      WRITE OUTPUT TAPE OUTPUT,2000,IDENT
      WRITE OUTPUT TAPE OUTPUT,2001
      DELX=1.0E-5
      ITERSTOP=100
      NRST=IR*JS*KT
      DO 10 N=1,NRST
      EXP(N)=OBS(N)
10 CONTINUE
      IR1=IR-1
      JS1=JS-1
      NX=IR1*JS1*KT
      ITERS=0
20 NXZERO=0
      ITERS=ITERS+1
      DO 60 I=1,IR1
      DO 50 J=1,JS1
      SUMC=0.0
      SUMBC=0.0
      DO 30 K=1,KT
      KPART=IR*(JS*K-JS-1)
      IJSK=KPART+IR*JS+I
      EXPISK=EXP(IJSK)
      IRJK=KPART+IR*J+I
      EXPRJK=EXP(IRJK)
      IRJSK=KPART+IR*JS+IR
      EXPRSK=EXP(IRJSK)
      IJK=KPART+IR*J+I
      EXPijk=EXP(IJK)
      CIJK=1.0/EXPISK+1.0/EXPRJK+1.0/EXPJSK+1.0/EXPijk
      CIJK=1.0/CIJK
      BIJK=EXPISK*EXPRJK*EXPJSK*EXPijk

```

```

SUMC=SUMC+CIJK
SUMBC=SUMBC+CIJK*BIJK
CIJ(K)=CIJK
BIJ(K)=BIJK
30 CONTINUE
HIJ=SUMC/SUMBC
DO 40 K=1,KT
KPART=IR*(JS*K-JS-1)
XIJK=CIJ(K)*(1.0-HIJ*BIJ(K))
IF(XIJK.LE.DELX) NXZERO=NXZERO+1
IRJSK=KPART+IR*JS+IR
EXP(IRJSK)=EXP(IRJSK)-XIJK
IJK=KPART+IR*J+I
EXP(IJK)=EXP(IJK)-XIJK
IJSK=KPART+IR*JS+I
EXP(IJSK)=EXP(IJSK)+XIJK
IRJK=KPART+IR*J+IR
EXP(IRJK)=EXP(IRJK)+XIJK
40 CONTINUE
50 CONTINUE
60 CONTINUE
IF(ITERSTOP.GE.ITERSTOP) GO TO 200
IF(NXZERO.LT.NX) GO TO 20
CHISQ=0.0
DO 70 N=1,NRST
EXPN=EXP(N)
CHISQ=CHISQ+(OBS(N)-EXPN)*(OBS(N)-EXPN)/EXPN
70 CONTINUE
WRITE OUTPUT TAPE OUTPUT,2003
IJK=0
DO 80 K=1,KT
DO 80 J=1,JS
DO 80 I=1,IR
IJK=IJK+1
WRITE OUTPUT TAPE OUTPUT,2005,I,J,K,OBS(IJK),EXP(IJK)
80 CONTINUE
WRITE OUTPUT TAPE OUTPUT,2002
WRITE OUTPUT TAPE OUTPUT,2004,ITERSTOP,NDEGREES,CHISQ,IR,JS,KT
GO TO 300
200 WRITE OUTPUT TAPE OUTPUT,2006,ITERSTOP
300 END

```

```

SUBROUTINE ARROCH
COMMON/ALL/OBS(400),IDENT(10),IR,JS,KT,NDEGREES,INPUT,OUTPUT
COMMON/SUMS/OBSR(80),OBSS(80),OBST(25),OBSRS(16),OBSTR(5),
*   OBSST(5),OBRSRT
COMMON/FLAGS/OPTION(10),LOGFLAG
DIMENSION EXP(400),THETA(80),PHI(80),PSI(25)
TYPE INTEGER OPTION
TYPE INTEGER OUTPUT
2000 FORMAT(1H1,10A8)
2001 FORMAT(19H0DARROCH PROCEDURE )
2002 FORMAT(///,12H ITERATION ,8X,18HDEGREES OF FREEDOM,8X,
*   14H CHI-SQUARE,15X,2H R,8X,2H S,8X,2H T)
2003 FORMAT(/////,8X,13H CELL ,8X,18HUSERVED FREQUENCY,8X,
*   18HEXPECTED FREQUFNCE /,1H )
2004 FORMAT(/3X,I4,18X,I5,15X,F15.6,7X,3(8X,I2))
2005 FORMAT(6X,3(2X,I3),8X,F15.6,11X,F15.6)
2006 FORMAT(25H0ITERATION COUNT EXCEEDS I3)
      WRITE OUTPUT TAPE OUTPUT,2000,IDENT
      WRITE OUTPUT TAPE OUTPUT,2001
      ITERSTOP=100
      TOLERSQ=0.00005*0.00005
      CALL GETSUMS
C
      DO 140 I=1,IR
      OBSSTI=OBSSST(I)
      DO 120 J=1,JS
      IJ=I+IR*(J-1)
      PSI(IJ)=OBST(IJ)/OBSSTI
120  CONTINUE
      DO 130 K=1,KT
      KI=K+KT*(I-1)
      PHI(KI)=OBSS(KI)/OBSRS(K)
130  CONTINUE
140  CONTINUE
C
      ITERS=0
200  ISTOP=1
      ITERS=ITERS+1
C
      DO 230 K=1,KT
      DO 220 J=1,JS
      JK=J+JS*(K-1)
      SUMI=0.0
      DO 210 I=1,IR
      KI=K+KT*(I-1)
      IJ=I+IR*(J-1)
      SUMI=SUMI+PHI(KI)*PSI(IJ)
210  CONTINUE
      THETAJK=OBSR(JK)/OBRSRT/SUVI
      IF((THETAJK-THETA(JK))**2.GE.TOLERSQ) ISTOP=0
      THETA(JK)=THETAJK
220  CONTINUE
230  CONTINUE

```

```

C
DO 260 I=1,IR
DO 250 K=1,KT
KI=K+KT*(I-1)
SUMJ=0.0
DO 240 J=1,JS
IJ=I+IR*(J-1)
JK=J+JS*(K-1)
SUMJ=SUMJ+PSI(IJ)*THETA(JK)
240 CONTINUE
PHIKI=OBSS(KI)/OBSRST/SUMJ
IF((PHIKI-PHI(KI))**2.GE.TOLERSQ) ISTOP=0
PHI(KI)=PHIKI
250 CONTINUE
260 CONTINUE

C
DO 290 J=1,JS
DO 280 I=1,IR
IJ=I+IR*(J-1)
SUMK=0.0
DO 270 K=1,KT
JK=J+JS*(K-1)
KI=K+KT*(I-1)
SUMK=SUMK+THETA(JK)*PHI(KI)
270 CONTINUE
PSIIJ=OBST(IJ)/OBSRST/SUMK
IF((PSIIJ-PSI(IJ))**2.GE.TOLERSQ) ISTOP=0
PSI(IJ)=PSIIJ
280 CONTINUE
290 CONTINUE
IF(ITERSTOP.GE.ITERSTOP) GO TO 400
IF(ISTOP.EQ.0) GO TO 200
IF(OPTION(4).EQ.0) GO TO 295
WRITE OUTPUT TAPE OUTPUT,2003
295 CHISQ=0.0
DO 300 K=1,KT
DO 300 J=1,JS
JK=J+JS*(K-1)
DO 300 I=1,IR
IJ=I+IR*(J-1)
KI=K+KT*(I-1)
EXPIJK=THETA(JK)*PHI(KI)*PSI(IJ)*OBSRST
IJK=I+IR*(J-1+JS*(K-1))
OBSSJK=OBSS(IJK)
CHISQ=CHISQ+(OBSSJK-EXPIJK)*(OBSSJK-EXPIJK)/EXPIJK
IF(OPTION(4).EQ.0) GO TO 300
WRITE OUTPUT TAPE OUTPUT,2005,I,J,K,OBSSJK,EXPIJK
300 CONTINUE
WRITE OUTPUT TAPE OUTPUT,2002
WRITE OUTPUT TAPE OUTPUT,2004,ITERS,NDEGREES,CHISQ,IR,JS,KT
GO TO 500

C
400 WRITE OUTPUT TAPE OUTPUT,2006,ITERSTOP
500 END

```

```

SUBROUTINE PLACKETT
COMMON/ALL/OBS(400),IDENT(10),IR,JS,KT,NDEGREES,INPUT,OUTPUT
DIMENSION V(256),VINV(256),L(16),ZINV(16)
TYPE INTEGER OUTPUT
2000 FORMAT(1H1,10A8)
2001 FORMAT(20H0PLACKETT PROCEDURE )
2002 FORMAT(////19H DEGREES OF FREEDOM ,8X,14H      CHI-SQUARE,17X,
* 10H SUM(L*L) ,15X,2H R,8X,2H S,8X,2H T)
2003 FORMAT(/6X,I5,15X,F15.6,11X,F15.6,8X,3(8X,I2))
2004 FORMAT(3H0K= I2)
2005 FORMAT(/12H    CONTRASTS )
2006 FORMAT(/20H    DISPERSION MATRIX )
2007 FORMAT(////11H    Z-VECTOR )
2008 FORMAT(/    11H    P-MATRIX )
WRITE OUTPUT TAPE OUTPUT,2000,IDENT
WRITE OUTPUT TAPE OUTPUT,2001
IR1=IR-1
JS1=JS-1
IR1JS1=IR1*JS1
ZINVZ=0.0
DO 30 NAB=1,IR1JS1
ZINV(NAB)=0.0
DO 30 NCD=1,IR1JS1
NABNCD=NAB+IR1JS1*(NCD-1)
VINV(NABNCD)=0.0
30 CONTINUE
DO 200 K=1,KT
DO 40 NAB=1,IR1JS1
Z(NAB)=0.0
DO 40 NCD=1,IR1JS1
NABNCD=NAB+IR1JS1*(NCD-1)
V(NABNCD)=0.0
40 CONTINUE
DO 170 J=1,JS
DO 170 I=1,IR
IJK=I+IR*(J-1+JS*(K-1))
OBSIJK=OBS(IJK)
OBSLOG=LOGF(OBSIJK)
DO 160 NB=1,JS1
SIGMA=1.0
IF(J+NB-JS-1) 60,50,170
50 SIGMA=-FLOATF(JS-NB)
60 DO 150 NA=1,IR1
RHO=1.0
IF(I+NA-IR-1) 80,70,160
70 RHO=-FLOATF(IR-NA)
80 NAB=IR-NA+IR1*(JS1-NB)
Z(NAB)=Z(NAB)+RHO*SIGMA*OBSLOG
DO 140 ND=1,JS1
OMEGA=1.0

```

```

      IF(J+ND-JS-1) 100,90,150
90  OMEGA=-FLQATE(JS-ID)
100 DO 130 NC=1,IR1
      TAU=1.0
      IF(1+NC-IR-1) 120,110,140
110  TAU=-FLQATE(IR-NC)
120  NCD=IR-NC+IR1*(JS1-ND)
      NABNCD=NAB+IR1JS1*(NCD-1)
      V(NABNCD)=V(NABNCD)+RH0*SIGMA*TAU*OMEGA/UBSIJK
130 CONTINUE
140 CONTINUE
150 CONTINUE
160 CONTINUE
170 CONTINUE
      WRITE OUTPUT TAPE OUTPUT,2004,K
      WRITE OUTPUT TAPE OUTPUT,2005
      CALL NMPRINT(Z,IR1JS1,1)
      WRITE OUTPUT TAPE OUTPUT,2006
      CALL NMPRINT(V,IR1JS1,IR1JS1)
      CALL MATINV(V,IR1JS1,V,0,D,IR1JS1)
      DO 190 NCD=1,IR1JS1
      ZINVCD=ZINV(NCD)
      ZCD=Z(NCD)
      DO 180 NAB=1,IR1JS1
      ZAB=Z(NAB)
      NABNCD=NAB+IR1JS1*(NCD-1)
      VABCD=V(NABNCD)
      ZINVZ=ZINVZ+ZAB*VABCD*ZCD
      VINV(NABNCD)=VINV(NABNCD)+VABCD
      ZINVCD=ZINVCD+ZAB*VABCD
180 CONTINUE
      ZINV(NCD)=ZINVCD
190 CONTINUE
200 CONTINUE
      CALL MATINV(VINV,IR1JS1,V,0,D,IR1JS1)
      CHISQ=ZINVZ
      DO 210 NCD=1,IR1JS1
      ZINVCD=ZINV(NCD)
      DO 210 NAB=1,IR1JS1
      NABNCD=NAB+IR1JS1*(NCD-1)
      CHISQ=CHISQ-ZINV(NAB)*VINV(NABNCD)*ZINVCD
210 CONTINUE
      WRITE OUTPUT TAPE OUTPUT,2007
      CALL NMPRINT(ZINV,IR1JS1,1)
      WRITE OUTPUT TAPE OUTPUT,2008
      CALL NMPRINT(VINV,IR1JS1,IR1JS1)
      WRITE OUTPUT TAPE OUTPUT,2002
      WRITE OUTPUT TAPE OUTPUT,2003,INDEGREES,CHISQ,ZINVZ,IR,JS,KT
300 END

```

```

SUBROUTINE GOODMAN1
COMMON/ALL/OBS(400),IDENT(10),IR,JS,KT,NDEGREES,INPUT,OUTPUT
DIMENSION V(256),VINV(256),Z(16),ZINV(16)
TYPE INTEGER OUTPUT
2000 FORMAT(1H1,10A8)
2001 FORMAT(19H0GOODMAN PROCEDURE )
2002 FORMAT(////19H DEGREES OF FREEDOM ,8X,14H      CHI-SQUARE,17X,
* 10H SUM(H*H) ,15X,2H R,8X,2H S,8X,2H T)
2003 FORMAT(/6X,I5,15X,F15.6,11X,F15.6,8X,3(8X,I2))
2004 FORMAT(3HOK= I2)
2005 FORMAT(/12H    CONTRASTS )
2006 FORMAT(/2UH    DISPERSION MATRIX )
2007 FORMAT(////11H    G-VECTOR )
2008 FORMAT(/    11H    Q-MATRIX )
      WRITE OUTPUT TAPE OUTPUT,2000,IDENT
      WRITE OUTPUT TAPE OUTPUT,2001
      IR1=IR-1
      JS1=JS-1
      IR1JS1=IR1*JS1
      ZINVZ=0.0
      DO 30 NAB=1,IR1JS1
      ZINV(NAB)=0.0
      DO 30 NCD=1,IR1JS1
      NABNCD=NAB+IR1JS1*(NCD-1)
      VINV(NABNCD)=0.0
30 CONTINUE
      DO 200 K=1,KT
      DO 40 NAB=1,IR1JS1
      Z(NAB)=0.0
      DO 40 NCD=1,IR1JS1
      NABNCD=NAB+IR1JS1*(NCD-1)
      V(NABNCD)=0.0
40 CONTINUE
      DO 170 J=1,JS
      DO 170 I=1,IR
      IJK=I+IR*(J-1+JS*(K-1))
      OBSIJK=OBS(IJK)
      OBSLOG=LOGF(OBSIJK)
      DO 160 NB=1,JS1
      SIGMA=-1.0
      IF(J.EQ.JS) GO TO 60
      IF(NB.NE.J) GO TO 160
      SIGMA=1.0
60 DO 150 NA=1,IR1
      RHO=-1.0
      IF(I.EQ.IR) GO TO 80
      IF(NA.NE.I) GO TO 150
      RHO=1.0
80 NAB=NA+IR1*(NB-1)
      Z(NAB)=Z(NAB)+RHO*SIGMA*OBSLOG

```

```

DO 140 ND=1,JS1
OMEGA=-1.0
IF(J.EQ.JS) GO TO 100
IF(ND.NE.J) GO TO 140
OMEGA=1.0
100 DO 130 NC=1,IR1
TAU=-1.0
IF(I.EQ.IR) GO TO 120
IF(NC.NE.I) GO TO 130
TAU=1.0
120 NCD=NC+IR1*(ND-1)
NABNCD=NAB+IR1JS1*(NCD-1)
V(NABNCD)=V(NABNCD)+RH0*SIGMA*TAU*CMEGA/CBSIJK
130 CONTINUE
140 CONTINUE
150 CONTINUE
160 CONTINUE
170 CONTINUE
WRITE OUTPUT TAPE OUTPUT,2004,K
WRITE OUTPUT TAPE OUTPUT,2005
CALL NMPRINT(Z,IR1JS1,1)
WRITE OUTPUT TAPE OUTPUT,2006
CALL NMPRINT(V,IR1JS1,IR1JS1)
CALL MATINV(V,IR1JS1,V,0,D,IR1JS1)
DO 190 NCD=1,IR1JS1
ZINVCD=ZINV(NCD)
ZCD=Z(NCD)
DO 180 NAB=1,IR1JS1
ZAB=Z(NAB)
NABNCD=NAB+IR1JS1*(NCD-1)
VABCD=V(NABNCD)
ZINVZ=ZINVZ+ZAB*VABCD*ZCD
VINV(NABNCD)=VINV(NABNCD)+VABCD
ZINVCD=ZINVCD+ZAB*VABCD
180 CONTINUE
ZINV(NCD)=ZINVCD
190 CONTINUE
200 CONTINUE
CALL MATINV(VINV,IR1JS1,V,0,D,IR1JS1)
CHISQ=ZINVZ
DO 210 NCD=1,IR1JS1
ZINVCD=ZINV(NCD)
DO 210 NAB=1,IR1JS1
NABNCD=NAB+IR1JS1*(NCD-1)
CHISQ=CHISQ-ZINV(NAB)*VINV(NABNCD)*ZINVCD
210 CONTINUE
WRITE OUTPUT TAPE OUTPUT,2007
CALL NMPRINT(ZINV,IR1JS1,1)
WRITE OUTPUT TAPE OUTPUT,2008
CALL NMPRINT(VINV,IR1JS1,IR1JS1)
WRITE OUTPUT TAPE OUTPUT,2009
WRITE OUTPUT TAPE OUTPUT,2003,NDEGREES,CHISQ,ZINVZ,TR,JC,KT
300 END

```

```

SUBROUTINE GOODMAN2
COMMON/ALL/OBS(400),IDENT(10),IR,JS,KT,NDEGREES,INPUT,OUTPUT
DIMENSION P(256),Q(256),B(80),C(16),G(16),PG(16)
TYPE INTEGER OUTPUT
2000 FORMAT(1H1,10A8)
2001 FORMAT(28HMODIFIED GOODMAN PROCEDURE ,//1H )
2002 FORMAT(/12H CONTRASTS )
2003 FORMAT(////19H DEGREES OF FREEDOM ,8X,14H CHI-SQUARE,15X,2H F
*   8X,2H S,8X,2H T )
2004 FORMAT(/6X,I5,15X,F15.6,7X,3(8X,I2))
      WRITE OUTPUT TAPE OUTPUT,2000,IDENT
      WRITE OUTPUT TAPE OUTPUT,2001
      IR1=IR-1
      JS1=JS-1
      IR1IR1=IR1*IR1
      IR1JS1=IR1*JS1
      CHISQ=0.0
      DO 15 NAB=1,IR1JS1
      PG(NAB)=0.0
      DO 10 NCD=1,IR1JS1
      NABNCD=NAB+IR1JS1*(NCD-1)
      Q(NABNCD)=0.0
10    CONTINUE
15    CONTINUE
      DO 200 K=1,KT
      DO 25 NAB=1,IR1JS1
      G(NAB)=0.0
      DO 20 NCD=1,IR1JS1
      NABNCD=NAB+IR1JS1*(NCD-1)
      P(NABNCD)=0.0
20    CONTINUE
25    CONTINUE
      DO 35 II=1,IR1
      DO 30 III=1,IR1
      IIIII=III+IR1*(II-1)
      C(IIII)=0.0
30    CONTINUE
35    CONTINUE
      DO 100 J=1,JS
      JK=IR*(J-1+JS*(K-1))
      OBSJK=0.0
      DO 70 I=1,IR
      IJK=I+JK
      OBSIJK=OBS(IJK)
      OBSLOG=LOGF(OBSIJK)
      DO 60 NB=1,JS1
      SIGMA=-1.0
      IF(J.EQ.JS) GO TO 40
      IF(NB.NE.J) GO TO 60
      SIGMA=1.0

```

```

40 DO 50 NA=1,IR1
RHO=-1.0
IF(I.EQ.IR1) GO TO 45
IF(NA.NE.I) GO TO 50
RHO=1.0
45 NAB=NA+IR1*(NB-1)
G(NAB)=G(NAB)+RHO*SIGMA*OBSLOG
50 CONTINUE
60 CONTINUE
OBSJK=OBSJK+OBSIJK
70 CONTINUE
DO 90 II=1,IR1
IIJK=II+JK
DO 80 III=1,IR1
IIIJK=III+JK
IIII=III+IR1*(II-1)
IIIIJ=IIII+IR1IR1*(J-1)
BIIIIJ=-OBS(IIJK)*OBS(IIIJK)/OBSJK
IF(II.EQ.III) BIIIIJ=BIIIIJ+OBS(IIJK)
C(IIII)=C(IIII)+BIIIIJ
B(IIIIJ)=BIIIIJ
80 CONTINUE
90 CONTINUE
100 CONTINUE
WRITE OUTPUT TAPE OUTPUT,2002
CALL NMPRINT(G,IR1JS1,1)
CALL MATINV(C,IR1,C,0,D,IR1)
DO 170 NB=1,JS1
DO 160 ND=1,JS1
DO 150 NA=1,IR1
NAB=NA+IR1*(NB-1)
DO 140 NC=1,IR1
NCD=NC+IR1*(ND-1)
NABNCD=NAB+IR1JS1*(NCD-1)
DO 120 II=1,IR1
IINCND=II+IR1*(NC-1+IR1*(ND-1))
DO 110 III=1,IR1
NAIINB=NA+IR1*(III-1+IR1*(NB-1))
IIII=III+IR1*(II-1)
P(NABNCD)=P(NABNCD)-B(NAIINB)*C(IIII)*B(IINCND)
110 CONTINUE
120 CONTINUE
IF(NB.NE.ND) GO TO 130
NANCND=NA+IR1*(NC-1+IR1*(ND-1))
P(NABNCD)=P(NABNCD)+B(NANCND)
130 G(NABNCD)=G(NABNCD)+P(NABNCD)
PG(NAB)=PG(NAB)+P(NABNCD)*G(NCD)
140 CONTINUE
150 CONTINUE
160 CONTINUE

```

```
170 CONTINUE
    DO 190 NAB=1,IR1JS1
    DO 180 NCD=1,IR1JS1
        NABNCD=NAB+IR1JS1*(NCD-1)
        CHISQ=CHISQ+G(NAB)*P(NABNCD)*G(NCD)
180 CONTINUE
190 CONTINUE
200 CONTINUE
    CALL MATINV(Q,IR1JS1,Q,0,D,IR1JS1)
    DO 220,NAB=1,IR1JS1
    DO 210 NCD=1,IR1JS1
        NABNCD=NAB+IR1JS1*(NCD-1)
        CHISQ=CHISQ-PG(NAB)*Q(NABNCD)*PG(NCD)
210 CONTINUE
220 CONTINUE
    WRITE OUTPUT TAPE OUTPUT,2003
    WRITE OUTPUT TAPE OUTPUT,2004,NDEGREES,CHISQ,IR,JS,KT
300 END
```

```

SUBROUTINE TWOBYTWO
COMMON/ALL/OBS(400),IDENT(10),IR,JS,KT,NDEGREES,INPUT,OUTPUT
DIMENSION U(16),V(16),G(16),U(16),S(16),A(16),DSTAR(16),VSTAR(16),
* GSTAR(16),USTAR(16),BSTAR(16),ASTAR(16)
TYPE INTEGER OUTPUT
2000 FORMAT(1H1,10A8)
2001 FORMAT(26H0GOODMANS 2X2XT PROCEDURE ,//1H )
2002 FORMAT(3H0 K,8X,3H D ,10X,3H V ,10X,3HX*X,10X,3H B ,10X,3H A ,10X,
* 3HZ*Z,10X,3H G ,10X,3H U ,10X,3HY*Y ,/1H )
2003 FORMAT(I3,9F13.5)
2004 FORMAT( I3,3(2F13.5,13X))
2005 FORMAT(//19H DEGREES OF FREEDOM ,11X,3HX*X,15X,3HZ*Z,15X,3HY*Y,
* 15X,2H R,8X,2H S,8X,2H T,/1H )
2006 FORMAT(6X,I5,6X,3F18.0,5X,3(8X,I2))
2007 FORMAT(3X,3(26X,F13.5))
      WRITE OUTPUT TAPE OUTPUT,2000,IDENT
      WRITE OUTPUT TAPE OUTPUT,2001
      IF(IR.NE.2) GO TO 100
      IF(JS.NE.2) GO TO 100
      WRITE OUTPUT TAPE OUTPUT,2002
      SUM1=SUM2=SUM3=SUM4=SUM5=SUM6=SUM7=SUM8=SUM9=0.0
      DO 20 K=1,KT
      KPART=IR*JS*(K-1)
      OBS11K=OBS(1+KPART)
      OBS21K=OBS(2+KPART)
      OBS12K=OBS(3+KPART)
      OBS22K=OBS(4+KPART)
      DK=OBS11K*OBS22K/OBS12K/OBS21K
      UK=1.0/OBS11K+1.0/OBS22K+1.0/OBS12K+1.0/OBS21K
      VK=DK*DK*UK
      WK=1.0/VK
      GK=LOGF(DK)
      PK=1.0/UK
      BK=1.0/DK
      AK=BK*BK*UK
      HK=1.0/AK
      SUM1=SUM1+DK*DK*WK
      SUM2=SUM2+DK*WK
      SUM3=SUM3+WK
      SUM4=SUM4+GK*GK*PK
      SUM5=SUM5+GK*PK
      SUM6=SUM6+PK
      SUM7=SUM7+BK*BK*HK
      SUM8=SUM8+BK*HK
      SUM9=SUM9+HK
      DSTAR(K)=SUM2/SUM3
      VSTAR(K)=1.0/SUM3

```

```

GSTAR(K)=SUM5/SUM6
USTAR(K)=1.0/SUM6
BSTAR(K)=SUM8/SUM9
ASTAR(K)=1.0/SUM9
D(K)=DK
V(K)=VK
G(K)=GK
U(K)=UK
B(K)=BK
A(K)=AK
20 CONTINUE
SUMXSQ=SUMYSQ=SUMZSQ=0.0
KT1=KT-1
DO 30 K=1,KT1
XSQK=(D(K+1)-DSTAR(K))*(D(K+1)-DSTAR(K))/(V(K+1)+VSTAR(K))
YSQK=(G(K+1)-GSTAR(K))*(G(K+1)-GSTAR(K))/(U(K+1)+USTAR(K))
ZSQK=(B(K+1)-BSTAR(K))*(B(K+1)-BSTAR(K))/(A(K+1)+ASTAR(K))
WRITE OUTPUT TAPE OUTPUT,2003,K,D(K),V(K),XSQK,B(K),A(K),ZSQK,
* G(K),U(K),YSQK
SUMXSQ=SUMXSQ+XSQK
SUMYSQ=SUMYSQ+YSQK
SUMZSQ=SUMZSQ+ZSQK
30 CONTINUE
WRITE OUTPUT TAPE OUTPUT,2004,KT,D(KT),V(KT),B(KT),A(KT),G(KT),
* U(KT)
WRITE OUTPUT TAPE OUTPUT,2007,SUMXSQ,SUMZSQ,SUMYSQ
XXSQ=SUM1-SUM2*SUM2/SUM3
YYSQ=SUM4-SUM5*SUM5/SUM6
ZZSQ=SUM7-SUM8*SUM8/SUM9
WRITE OUTPUT TAPE OUTPUT,2005
WRITE OUTPUT TAPE OUTPUT,2006,NDEGREES,XXSQ,ZZSQ,YYSQ,IR,JS,KT
100 END

```

```

SUBROUTINE KKK
COMMON/ALL/OBS(400),IDENT(10),IR,JS,KT,NDEGREES,INPUT,OUTPUT
COMMON/SUMS/OBSR(80),OBSS(80),OBST(25),OBSRS(16),CRSTR(5),
*   OBSST(5),OBSRST
TYPE INTEGER OUTPUT
2000 FORMAT(1H1,10A8)
2001 FORMAT(39HOKULLBACK, KUPPERMAN, AND KU PROCEDURE )
2002 FORMAT(////19H DEGREES OF FREEDOM ,11X,3H2*I,17X,2H R,8X,2H S,8X,
* 2H T)
2003 FORMAT(/6X,I5,6X,F18.5,7X,3(8X,I2))
      WRITE OUTPUT TAPE OUTPUT,2000,IDENT
      WRITE OUTPUT TAPE OUTPUT,2001
      CALL GETSUMS
      TWOI=0.0
      DO 140 K=1,KT
      OBSRSK=OBSRS(K)
      DO 130 J=1,JS
      OBSTRJ=OBSTR(J)
      JK=J+JS*(K-1)
      OBSRJK=OBSR(JK)
      DO 120 I=1,IR
      OBSSTI=OBSST(I)
      IJ=I+IR*(J-1)
      OBSTIJ=OBST(IJ)
      KI=K+KT*(I-1)
      OBSSKI=OBSS(KI)
      IJK=I+IR*(J-1+JS*(K-1))
      OBSIJK=OBS(IJK)
      TWOI=TWOI+OBSIJK*LOGF(OBSIJK*OBSSTI*OBSTRJ*OBSRSK/OBSRST/OBSTIJ/
*   OBSSKI/OBSRJK)
120 CONTINUE
130 CONTINUE
140 CONTINUE
      TWOI=2.0*TWOI
      WRITE OUTPUT TAPE OUTPUT,2002
      WRITE OUTPUT TAPE OUTPUT,2003,NDEGREES,TWOI,IR,JS,KT
400 END

```

```

SUBROUTINE INTRVALS
COMMON/BCD/IBVECTOR(20),JCVECTOR(20),KDVECTOR(20)
COMMON/ALL/OBS(400),IDENT(10),IR,JS,KT,NDEGREES,INPUT,OUTPUT
COMMON/FLAGS/OPTION(10),LOGFLAG
DIMENSION Q(6),CHIT(2),CHI1(2),PHIW(2),P(2)
TYPE INTEGER OUTPUT
TYPE INTEGER OPTION
DATA(P=0.95,0.99)
2000 FORMAT(1H1,10A8)
2001 FORMAT(61H0INTERACTION ESTIMATES AND CONFIDENCE INTERVALS IN LOG U
    *NITS )
2002 FORMAT(63H0INTERACTION ESTIMATES AND CONFIDENCE INTERVALS IN BASIC
    * UNITS )
2003 FORMAT(//// 44H          B           C           D   )
2004 FORMAT(/9X,I2,14X,I2,14X,I2)
2005 FORMAT(//// 34H          P           CHI(T=I5,43H)
    *      CHI(1)          PHI(W=I5,1H),/1H )
2006 FORMAT(9X,F4.2,3(9X,F15.4))
2007 FORMAT(/////106H0 I  B  J  C  K  D      P      ESTIMATE      S
    *STANDARD          CHI(T)          CHI(1)
    *      PHI(W) /,1H )
2008 FORMAT(6I3,F9.2,2E15.3,4X,3(4X,2E10.2))
2009 FORMAT(/////106H0 I  B  J  C  K  D      P      ESTIMATE
    *          CHI(1)          PHI(W) /,1H )
2010 FORMAT(6I3,F9.2,E15.3,15X,4X,3(4X,2E10.2))
      WRITE OUTPUT TAPE OUTPUT,2000,IDENT
      IF(LOGFLAG.EQ.0) GO TO 5
      WRITE OUTPUT TAPE OUTPUT,2001
      GO TO 10
      5 WRITE OUTPUT TAPE OUTPUT,2002
      10 WRITE OUTPUT TAPE OUTPUT,2003
      NW=0
      NBCD=0
      DO 15 I=1,20
      IB=IBVECTOR(I)
      IF(IB.EQ.0) GO TO 15
      JC=JCVECTOR(I)
      KD=KDVECTOR(I)
      NBCD=NBCD+1
      NW=NW+(IB-1)*(JC-1)*(KD-1)
      WRITE OUTPUT TAPE OUTPUT,2004,IB,JC,KD
15 CONTINUE
      WRITE OUTPUT TAPE OUTPUT,2005,NDEGREES,NW
      W=FLOATF(NW)
      T=FLOATF(NDEGREES)
      DO 20 L=1,2
      PL=P(L)
      CHITL=CHI(T,PL)

```

```

CHI1L=CHI(1,0,PL)
PHIWL=PHI(W,PL)
WRITE OUTPUT TAPE OUTPUT,2006,PL,CHITL,CHI1L,PHIWL
CHIT(L)=CHITL
CHI1(L)=CHI1L
PHIW(L)=PHIWL
20 CONTINUE
IF(LOGFLAG.EQ.0) GO TO 25
WRITE OUTPUT TAPE OUTPUT,2007
GO TO 30
25 WRITE OUTPUT TAPE OUTPUT,2009
30 DO 100 N=1,NBCD
KD=KDVVECTOR(N)
KD1=KD-1
JC=JCVECTOR(N)
JC1=JC-1
IB=IBVECTOR(N)
IB1=IB-1
DO 70 K=1,KD1
DO 60 J=1,JC1
DO 50 I=1,IB1
IJK=I+IR*(J-1+JS*(K-1))
OBSIJK=OBS(IJK)
IBJCK=IB+IR*(JC-1+JS*(K-1))
OBSBCK=OBS(IBJCK)
IBJK=IB+IR*(J-1+JS*(K-1))
OBSBJK=OBS(IBJK)
IJCK=I+IR*(JC-1+JS*(K-1))
OBSICK=OBS(IJCK)
IJKD=I+IR*(J-1+JS*(KD-1))
OBSIJD=OBS(IJKD)
IBJCKD=IB+IR*(JC-1+JS*(KD-1))
OBSBCD=OBS(IBJCKD)
IBJKD=IB+IR*(J-1+JS*(KD-1))
OBSBJD=OBS(IBJKD)
IJCKD=I+IR*(JC-1+JS*(KD-1))
OBSICD=OBS(IJCKD)
AIBJCKD=LOGF(OBSIJK*OBSBCK*OBSBJD*OBSICD/OBSBJK/OBSICK/OBSIJD/
*          OBSBCD)
UIBJCK=1.0/OBSIJK+1.0/UOBSBCK+1.0/UOBSICK+1.0/UOBSBJK
UIBJCD=1.0/OBSIJD+1.0/UOBSBCD+1.0/UOBSICD+1.0/UOBSBJD
SIBJCKD=SORTF(UIBJCK+UIBJCD)
DO 45 L=1,2
PL=P(L)
Q(1)=AIBJCKD-SIBJCKD*CHIT(L)
Q(2)=2.0*AIBJCKD-Q(1)
Q(3)=AIBJCKD-SIBJCKD*CHI1(L)
Q(4)=2.0*AIBJCKD-Q(3)
Q(5)=AIBJCKD-SIBJCKD*PHIW(L)

```

```
Q(6)=2.0*AIBJCKD-Q(5)
IF(LOGFLAG.EQ.0) GO TO 35
WRITE OUTPUT TAPE OUTPUT,2008,I,IB,J,JC,K,KD,PL,AIBJCKD,SIBJCKD,Q
GO TO 45
35 AIBJCKD=EXPF(AIBJCKD)
DO 40 M=1,6
Q(M)=EXPF(Q(M))
40 CONTINUE
WRITE OUTPUT TAPE OUTPUT,2010,I,IB,J,JC,K,KD,PL,AIBJCKD,Q
45 CONTINUE
50 CONTINUE
60 CONTINUE
70 CONTINUE
100 CONTINUE
END
```

```
SUBROUTINE READIN
COMMON/ALL/OBS(400),IDENT(10),IR,JS,KT,NDEGREES,INPUT,OUTPUT
COMMON/BCD/IBVECTOR(20),JCVECTOR(20),KDVECTOR(20)
COMMON/FLAGS/OPTION(10),LOGFLAG
TYPE INTEGER OPTION
TYPE INTEGER OUTPUT
DIMENSION INMAT(10)
1000 FORMAT(10A8)
1001 FORMAT(2U14)
1002 FORMAT(10I1)
      READ INPUT TAPE INPUT,1000,IDENT
      IF(IDENT(1).EQ.6HTHAT'S ALL) GO TO 400
      READ INPUT TAPE INPUT,1002,OPTION
      READ INPUT TAPE INPUT,1001,IR,JS,KT
      READ INPUT TAPE INPUT,1000,INMAT
      NRST=IR*JS*KT
      READ INPUT TAPE INPUT,INMAT,(OBS(N),N=1,NRST)
      READ INPUT TAPE INPUT,1001,IBVECTOR
      READ INPUT TAPE INPUT,1001,JCVECTOR
      READ INPUT TAPE INPUT,1001,KDVECTOR
      NDEGREES=(IR-1)*(JS-1)*(KT-1)
      RETURN
400 STOP
END
```

```

SUBROUTINE GETSUMS
COMMON/ALL/OBS(400),IDENT(10),IR,JS,KT,NDEGREES,INPUT,OUTPUT
COMMON/SUMS/OBSR(80),OBSS(80),OBST(25),OBSRS(16),OBSTR(5),
*   OBSST(5),OBSRST
OBSRST=0.0
DO 30 K=1,KT
OBSRSK=0.0
DO 20 J=1,JS
OBSRJK=0.0
DO 10 I=1,IR
IJK=I+IR*(J-1+JS*(K-1))
OBSIJK=OBS(IJK)
OBSRJK=OBSRJK+OBSIJK
OBSRSK=OBSRSK+OBSIJK
OBSRST=OBSRST+OBSIJK
10 CONTINUE
JK=J+JS*(K-1)
OBSR(JK)=OBSRJK
20 CONTINUE
OBSRS(K)=OBSRSK
30 CONTINUE
C
DO 60 J=1,JS
OBSTRJ=0.0
DO 50 I=1,IR
OBSTIJ=0.0
DO 40 K=1,KT
IJK=I+IR*(J-1+JS*(K-1))
OBSIJK=OBS(IJK)
OBSTIJ=OBSTIJ+OBSIJK
OBSTRJ=OBSTRJ+OBSIJK
40 CONTINUE
IJ=I+IR*(J-1)
OBST(IJ)=OBSTIJ
50 CONTINUE
OBSTR(J)=OBSTRJ
60 CONTINUE
C
DO 90 I=1,IR
OBSSTI=0.0
DO 80 K=1,KT
OBSSKI=0.0
DO 70 J=1,JS
IJK=I+IR*(J-1+JS*(K-1))
OBSIJK=OBS(IJK)
OBSSKI=OBSSKI+OBSIJK
OBSSTI=OBSSTI+OBSIJK
70 CONTINUE
KI=K+KT*(I-1)
OBSS(KI)=OBSSKI
80 CONTINUE
OBSST(I)=OBSSTI
90 CONTINUE
END

```

```

SUBROUTINE NMPRINT(X,N,M)
COMMON/ALL/UBS(400),IDENT(10),IR,JS,KT,NDEGREES,INPUT,OUTPUT
DIMENSION X(256),XNORM(256)
TYPE INTEGER OUTPUT
2000 FORMAT(9X,32HALL NUMBERS HAVE BEEN DIVIDED BY E11.1/,1H )
2001 FORMAT(8X,16F7.3)
NM=N*M
XMAX=X(1)
DO 10 K=2,NM
IF(XMAX*XMAX.GT.X(K)*X(K)) GO TO 10
XMAX=X(K)
10 CONTINUE
XMAX=ABSF(XMAX)
XMAXLOG=LOGF(XMAX)/2.3025850930
MAXLOG=XMAXLOG
IF(XMAX.LT.1.0) MAXLOG=MAXLOG-1
XMAX=10.0**MAXLOG
DO 20 K=1,NM
XNORM(K)=X(K)/XMAX
20 CONTINUE
WRITE OUTPUT TAPE OUTPUT,2000,XMAX
DO 30 I=1,M
J2=N*I
J1=J2-N+1
WRITE OUTPUT TAPE OUTPUT,2001,(XNORM(J),J=J1,J2)
30 CONTINUE
END

```

```

FUNCTION PHI(W,PL)
AREA=1.0+(PL-1.0)/2.0/W
TEST=0.
EP = .0000001
IF(AREA-.5)3,4,5
4 XORD=0.
GOTO16
3 AREA1=1.-AREA
C = -1.
GOTO7
5 AREA1=ARFA
C = 1.
7 X = (AREA1 - .5) * 2.5
8 DIV = 1. + .3275911 * X
E = 1.0/ DIV
S = ((((.940646070*E)-1.287822453)*E+1.259695130)*E-.252128668)*E
X+.225836846
SD = (((((4.70323035      * E ) - 5.151289812      )*E +
X3.77908539 ) * E - .504257336      ) * E + .225830846 ) * E
AA = EXPF ( X **2 ) * .88622692
FX = (AREA1 - 1. ) * AA * DIV * 2. + S
FXD = - SD * .3275911      - S * 2. * X
XNEW = X - FX / FXP
IF(ABSF (XNEW-X) -EP ) 6, 6, 14
14 TEST=TEST+1.
IF(TEST-40.)15,6,6
15 X = XNEW
GO TO 8
6 XORD = C * XNEW * 1.414213562
16 PHI=XORD
END

```

```
FUNCTION CHI(T,PL)
DIMENSION X(13,2)
DATA(X=3.641,5.991,7.015,9.400,11.070,1.000,2.32638131,1.15732,
* -0.554967955,-0.123011797,0.077939557,-0.100617878,0.122445162,
* 6.635,9.210,11.345,13.277,15.036,1.000,3.28946075,2.94018400,
* -0.290518574,-0.341655265,0.411541798,-0.342315613,0.20111198)
L=1
IF(PL.GT.0.975) L=2
IF(T.GT.5.5) GO TO 10
IT=T
CHI=X(IT,L)
GO TO 30
10 TRT=SQRTF(T)
CHI=X(13,L)
DO 20 K=1,7
CHI=CHI/TRT+X(13-K,L)
20 CONTINUE
CHI=T*CHI
30 CHI=SQRTF(CHI)
END
```

TABLE 1 - DATA FROM KASTENBAUM AND LAMPHIEAR (1959) - 2 X 3 X 5
KASTENBAUM-LAMPHIEAR PROCEDURE

| CELL | | | OBSERVED FREQUENCY | EXPECTED FREQUENCY |
|------|---|---|--------------------|--------------------|
| 1 | 1 | 1 | 58.000000 | 54.995024 |
| 2 | 1 | 1 | 75.000000 | 78.004976 |
| 1 | 2 | 1 | 11.000000 | 12.381343 |
| 2 | 2 | 1 | 19.000000 | 17.618657 |
| 1 | 2 | 1 | 5.000000 | 6.623632 |
| 2 | 3 | 1 | 7.000000 | 5.376368 |
| 1 | 1 | 2 | 49.000000 | 48.379034 |
| 2 | 1 | 2 | 58.000000 | 58.620966 |
| 1 | 2 | 2 | 14.000000 | 13.991513 |
| 2 | 2 | 2 | 17.000000 | 17.008487 |
| 1 | 3 | 2 | 10.000000 | 10.629453 |
| 2 | 3 | 2 | 8.000000 | 7.370547 |
| 1 | 1 | 3 | 33.000000 | 34.127160 |
| 2 | 1 | 3 | 45.000000 | 43.872840 |
| 1 | 2 | 3 | 18.000000 | 17.469260 |
| 2 | 2 | 3 | 22.000000 | 22.530740 |
| 1 | 3 | 3 | 15.000000 | 14.403580 |
| 2 | 3 | 3 | 10.000000 | 10.596420 |
| 1 | 1 | 4 | 15.000000 | 17.132082 |
| 2 | 1 | 4 | 39.000000 | 36.867918 |
| 1 | 2 | 4 | 13.000000 | 11.079608 |
| 2 | 2 | 4 | 22.000000 | 23.920392 |
| 1 | 3 | 4 | 15.000000 | 14.788311 |
| 2 | 3 | 4 | 18.000000 | 18.211689 |
| 1 | 1 | 5 | 4.000000 | 4.366700 |
| 2 | 1 | 5 | 5.000000 | 4.633300 |
| 1 | 2 | 5 | 12.000000 | 13.078277 |
| 2 | 2 | 5 | 15.000000 | 13.921723 |
| 1 | 3 | 5 | 17.000000 | 15.555023 |
| 2 | 3 | 5 | 8.000000 | 9.444977 |

| ITERATION | DEGREES OF FREEDOM | CHI-SQUARE |
|-----------|--------------------|------------|
| 26 | 8 | 3.158045 |

| R | S | T |
|---|---|---|
| 2 | 3 | 5 |

TABLF 2 - DATA FROM KASTENBAUM AND LAMPHIFAR (1959) - 2 X 3 X 5

PLACKETT PROCEDURE

K= 1

CONTRASTS

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0-001
2.895 -1.304

DISPERSION MATRIX

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0+000
0.174 -0.113
-0.113 1.546

K= 2

CONTRASTS

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0-001
0.255 -0.091

DISPERSION MATRIX

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0+000
0.158 -0.093
-0.093 1.068

K= 3

CONTRASTS

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0+000
-0.179 -1.322

DISPERSION MATRIX

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0-001
1.535 -0.485
-0.485 0.202

TABLE 2 (Continued)

K = 4

CONTRASTS

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0+000
 -0.429 -1.117

DISPERSION MATRIX

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0-001
 2.147 -0.301
 -0.301 7.036

K = 5

CONTRASTS

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0+000
 0.000 -1.954

DISPERSION MATRIX

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0+000
 0.600 0.300
 0.300 1.335

Z-VECTOR

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0+000
 -1.248 -5.760

P-MATRIX

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0-001
 0.336 -0.088
 -0.088 1.951

DEGREES OF FREEDOM

CHI-SQUARE

SUM(L*L)

8

3.127641

9.533446

| | | |
|---|---|---|
| R | S | T |
| 2 | 3 | 5 |

TABLE 3 - DATA FROM KASTENBAUM AND LAMPHIFAR (1959) - 2 X 3 X 5
GOODMAN PROCEDURE

K = 1

CONTRASTS

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0-001
0.794 -2.101

DISPERSION MATRIX

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0-001
3.734 3.429
3.429 4.954

K = 2

CONTRASTS

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0-001
-3.918 -4.173

DISPERSION MATRIX

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0-001
2.626 2.250
2.250 3.553

K = 3

CONTRASTS

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0-001
-7.156 -6.061

DISPERSION MATRIX

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0-001
2.192 1.667
1.667 2.677

TABLE 3 (Continued)

K= 4

CONTRASTS

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0-001
-7.732 -3.438

DISPERSION MATRIX

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0-001
2.145 1.222
1.222 2.446

K= 5

CONTRASTS

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0-001
-9.769 -9.769

DISPERSION MATRIX

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0-001
6.338 1.838
1.838 3.338

G-VECTOR

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0+000
-7.008 -4.511

Q-MATRIX

ALL NUMBERS HAVE BEEN DIVIDED BY 1.0-002
5.426 3.886
3.886 6.309

DEGREES OF FREEDOM

CHI-SQUARE

SUM(H*H)

8 3.127641

9.533446

| R | S | T |
|---|---|---|
| 2 | 3 | 5 |

TABLE 4 - DATA FROM NORTON (1945) - 2 X 2 X 12

GOODMANS 2X2XT PROCEDURE

| K | D | V | X**X | B | A |
|----|---------|---------|---------|----------|------------|
| 1 | 0.05543 | 0.00084 | 0.29711 | 18.03922 | 88.58704 |
| 2 | 0.03516 | 0.00041 | 0.02239 | 27.65714 | 242.38946 |
| 3 | 0.04712 | 0.00066 | 0.21299 | 21.22105 | 133.87251 |
| 4 | 0.06208 | 0.00136 | 0.96746 | 16.10714 | 91.40393 |
| 5 | 0.02715 | 0.00020 | 0.00713 | 36.83333 | 375.06906 |
| 6 | 0.03947 | 0.00045 | 0.03231 | 25.33333 | 184.97354 |
| 7 | 0.03393 | 0.00040 | 0.95887 | 29.47368 | 300.12538 |
| 8 | 0.07353 | 0.00131 | 0.92882 | 13.60000 | 44.85733 |
| 9 | 0.07045 | 0.00101 | 0.14199 | 14.19444 | 40.97770 |
| 10 | 0.03316 | 0.00035 | 1.69991 | 30.16000 | 285.59509 |
| 11 | 0.01931 | 0.00022 | 2.18260 | 54.62500 | 1940.89453 |
| 12 | 0.01655 | 0.00013 | | 60.41667 | 1695.44271 |
| | | | 7.45158 | | |

| Z*Z | G | U | Y*Y |
|---------|----------|---------|---------|
| 0.27949 | -2.89255 | 0.27223 | 0.30999 |
| 0.00186 | -3.31988 | 0.31688 | 0.00277 |
| 0.16383 | -3.05499 | 0.29727 | 0.19875 |
| 0.76060 | -2.77926 | 0.35231 | 0.99599 |
| 0.10675 | -3.60640 | 0.27640 | 0.02324 |
| 0.21198 | -3.23212 | 0.28822 | 0.12904 |
| 1.00329 | -3.38350 | 0.34549 | 1.16016 |
| 0.42970 | -2.61007 | 0.24252 | 0.82690 |
| 0.51606 | -2.65285 | 0.20338 | 0.41032 |
| 0.67826 | -3.40652 | 0.31397 | 1.29140 |
| 1.03293 | -4.00049 | 0.65046 | 2.02544 |
| | -4.10127 | 0.46448 | |
| 5.18475 | | | 7.37399 |

DEGREES OF FREEDOM

X**X

Z*Z

Y*Y

11 7.45158 5.18475 7.37399

| R | S | T |
|---|---|----|
| 2 | 2 | 12 |

TABLE 5 - DATA FROM KASTENBAUM AND LAMPHIFAR (1959) - 2 X 3 X 5
INTERACTION ESTIMATES AND CONFIDENCE INTERVALS IN LOG UNITS

| | | |
|---|---|---|
| R | C | D |
| 2 | 3 | 5 |

| P | CHI(T= 8) | CHI(1) |
|------|-----------|--------|
| 0.95 | 3.9381 | 1.9598 |
| 0.99 | 4.4819 | 2.5758 |

| I | R | J | C | K | N | P | FESTIMATE | STANDARD ERROR |
|---|---|---|---|---|---|------|-----------|-------------------|
| 1 | 2 | 1 | 3 | 1 | 5 | 0.95 | 1.056+000 | 1.004+000 |
| 1 | 2 | 1 | 3 | 1 | 5 | 0.99 | 1.056+000 | 1.004+000 |
| 1 | 2 | 2 | 3 | 1 | 5 | 0.95 | 7.668-001 | 9.057-001 |
| 1 | 2 | 2 | 3 | 1 | 5 | 0.99 | 7.668-001 | 9.057-001 |
| 1 | 2 | 1 | 3 | 2 | 5 | 0.95 | 5.851-001 | 9.468-001 |
| 1 | 2 | 1 | 3 | 2 | 5 | 0.99 | 5.851-001 | 9.468-001 |
| 1 | 2 | 2 | 3 | 2 | 5 | 0.95 | 5.596-001 | 8.301-001 |
| 1 | 2 | 2 | 3 | 2 | 5 | 0.99 | 5.596-001 | 8.301-001 |
| 1 | 2 | 1 | 3 | 3 | 5 | 0.95 | 2.613-001 | 9.236-001 |
| 1 | 2 | 1 | 3 | 3 | 5 | 0.99 | 2.613-001 | 9.236-001 |
| 1 | 2 | 2 | 3 | 3 | 5 | 0.95 | 3.708-001 | 7.756-001 |
| 1 | 2 | 2 | 3 | 3 | 5 | 0.99 | 3.708-001 | 7.756-001 |
| 1 | 2 | 1 | 3 | 4 | 5 | 0.95 | 2.037-001 | 9.211-001 |
| 1 | 2 | 1 | 3 | 4 | 5 | 0.99 | 2.037-001 | 9.211-001 |
| 1 | 2 | 2 | 3 | 4 | 5 | 0.95 | 6.331-001 | 7.505-001 |
| 1 | 2 | 2 | 3 | 4 | 5 | 0.99 | 6.331-001 | 7.605-001 |

TABLE 5 (Continued)

PHI(W= 8)

2.7344
3.2272

| CHI(T) | | ...LIMITS... | | PHI(W) |
|-----------|----------|--------------|----------|-------------------|
| | | CHI(1) | | |
| -2.90+000 | 5.01+000 | -9.11-001 | 3.02+000 | -1.69+000 3.80+00 |
| -3.44+000 | 5.55+000 | -1.53+000 | 3.64+000 | -2.18+000 4.30+00 |
| -2.80+000 | 4.33+000 | -1.01+000 | 2.54+000 | -1.71+000 3.24+00 |
| -3.29+000 | 4.83+000 | -1.57+000 | 3.10+000 | -2.16+000 3.69+00 |
| -3.14+000 | 4.31+000 | -1.27+000 | 2.44+000 | -2.00+000 3.17+00 |
| -3.66+000 | 4.83+000 | -1.85+000 | 3.02+000 | -2.47+000 3.64+00 |
| -2.71+000 | 3.83+000 | -1.07+000 | 2.19+000 | -1.71+000 2.83+00 |
| -3.16+000 | 4.28+000 | -1.58+000 | 2.70+000 | -2.12+000 3.24+00 |
| -3.38+000 | 3.90+000 | -1.55+000 | 2.07+000 | -2.26+000 2.79+00 |
| -3.88+000 | 4.40+000 | -2.12+000 | 2.64+000 | -2.72+000 3.24+00 |
| -2.68+000 | 3.42+000 | -1.15+000 | 1.89+000 | -1.75+000 2.49+00 |
| -3.11+000 | 3.85+000 | -1.63+000 | 2.37+000 | -2.13+000 2.87+00 |
| -3.42+000 | 3.83+000 | -1.60+000 | 2.01+000 | -2.31+000 2.72+00 |
| -3.92+000 | 4.33+000 | -2.17+000 | 2.58+000 | -2.77+000 3.18+00 |
| -2.36+000 | 3.63+000 | -8.57-001 | 2.12+000 | -1.45+000 2.71+00 |
| -2.78+000 | 4.04+000 | -1.33+000 | 2.59+000 | -1.82+000 3.09+00 |